

U.S. DEPARTMENT OF ENERGY

# **SMARTMOBILITY**

Systems and Modeling for Accelerated Research in Transportation

Systems and Modeling for Accelerated Research in Transportation (SMART) Mobility Consortium Tools and Process Development

Aymeric Rousseau Argonne National Laboratory 2019 Vehicle Technologies Office Annual Merit Review June 12, 2019











# **PROJECT OVERVIEW**

Timeline	Barriers		
<ul> <li>Project start date : Oct. 2018</li> <li>Project end date : Sep. 2019</li> <li>Percent complete : 60%</li> </ul>	<ul> <li>High uncertainty in technology deployment, functionality, usage, impact at system level</li> <li>Computational models, design and simulation methodologies</li> <li>Integration of many model frameworks: land use, demand, flow, vehicles, grid, economy</li> </ul>		
Budget	Partners		
• FY19 Funding Received : \$1,000,000	<ul> <li>Argonne (Lead)</li> <li>LBNL, NREL, ORNL, INL, LLNL</li> <li>Universities (UCI, GMU, UIC, Texas A&amp;M, Taxas At Austin, UNSW, Washington)</li> </ul>		













# PROJECT RELEVANCE

System level workflow is required to answer complex questions and provide actionable information

- What is the impact of vehicle fleet sharing, multi-modal travel, personally owned fully automated vehicles on mobility, energy, Vehicle Miles Travelled (VMT), Mobility Energy Productivity (MEP)...?
- How is intra-city freight impacted by disruptive technologies, such as ecommerce, electrification, in-route passenger delivery systems?
- What is the potential to increase efficiency through advanced vehicle control enabled by connectivity and automation?

Workflow needs to be deployed to engage with stakeholders and other researchers

- Computationally efficient (<12h)</li>
- Easy to use
- Deployable process
- Model agnostic





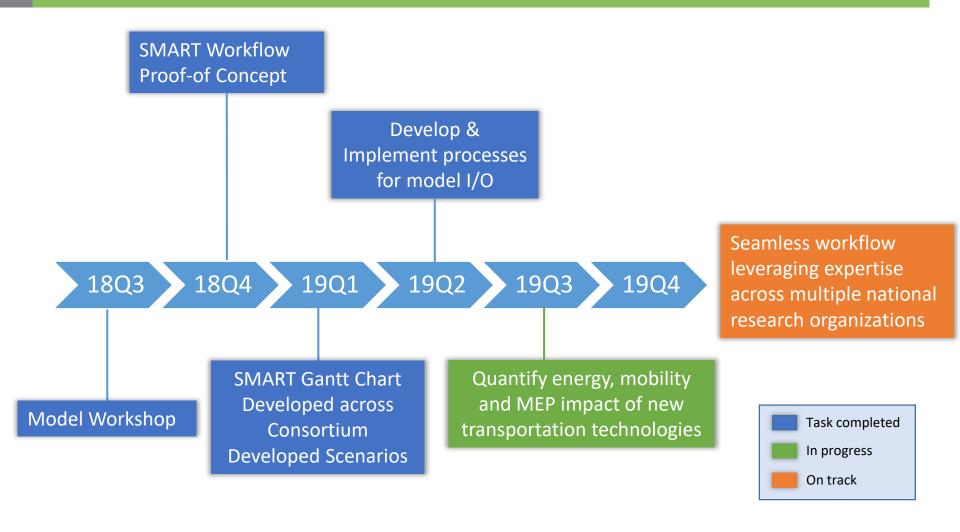








## **MILESTONES**







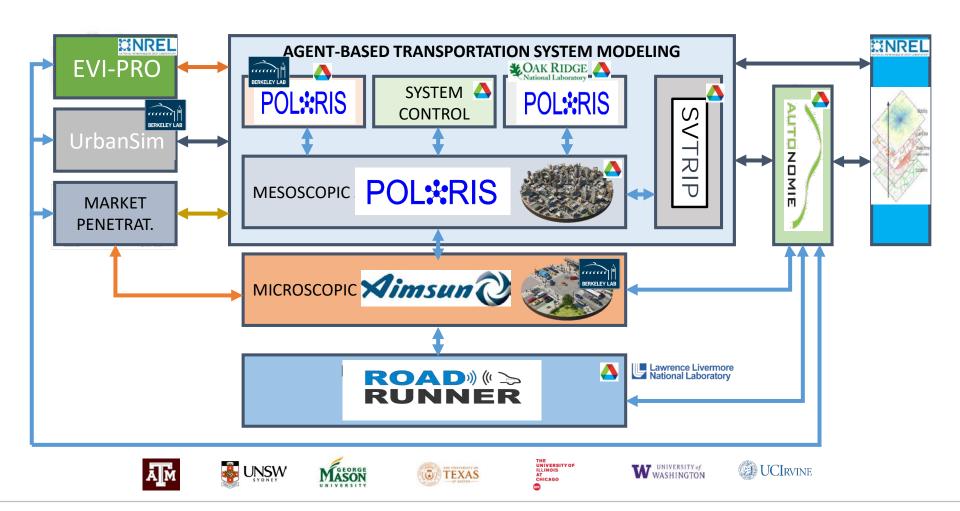








# APPROACH - SMART WORKFLOW A COMPREHENSIVE APPROACH TO ANSWER COMPLEX QUESTIONS























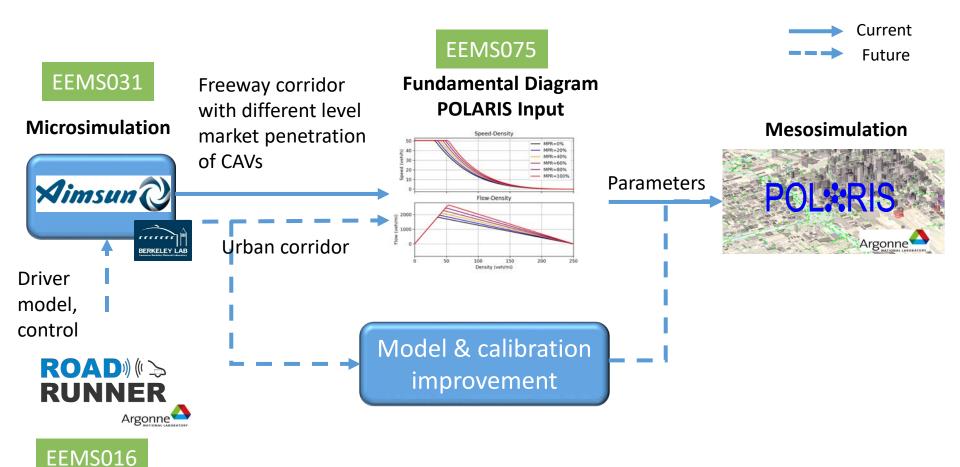






# LEARNING FROM DETAILED MODELS TO SCALE TO LARGER ONES











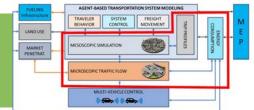


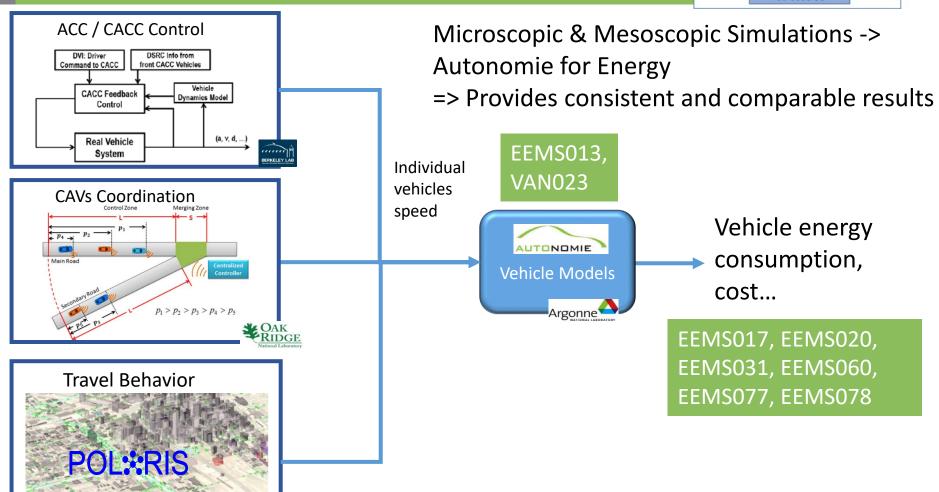




### **AUTONOMIE REUSED ACROSS CONSORTIUM**

Argonne™









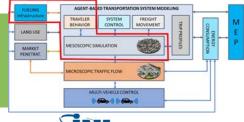




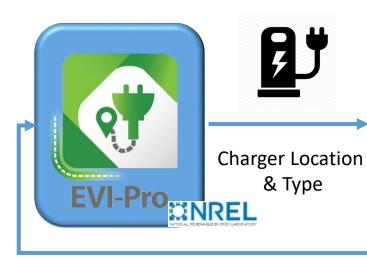




## PEV CHARGING LOCATION AND BEHAVIOR















Machine Learning for **Energy Consumption** 

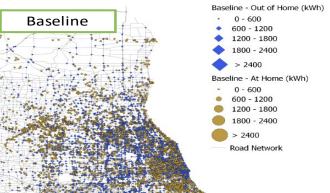
Scenario B

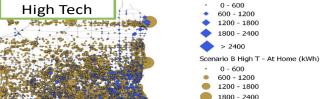


Charging Decision



EEMS068









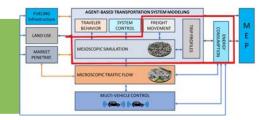


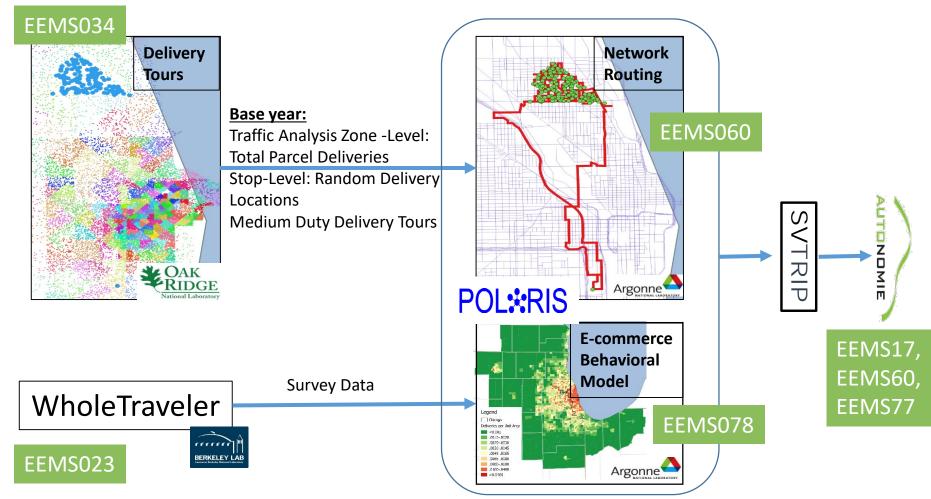


Scenario B High T - Out of Home (kWh)



## IMPROVED SCENARIOS – FREIGHT EXAMPLE









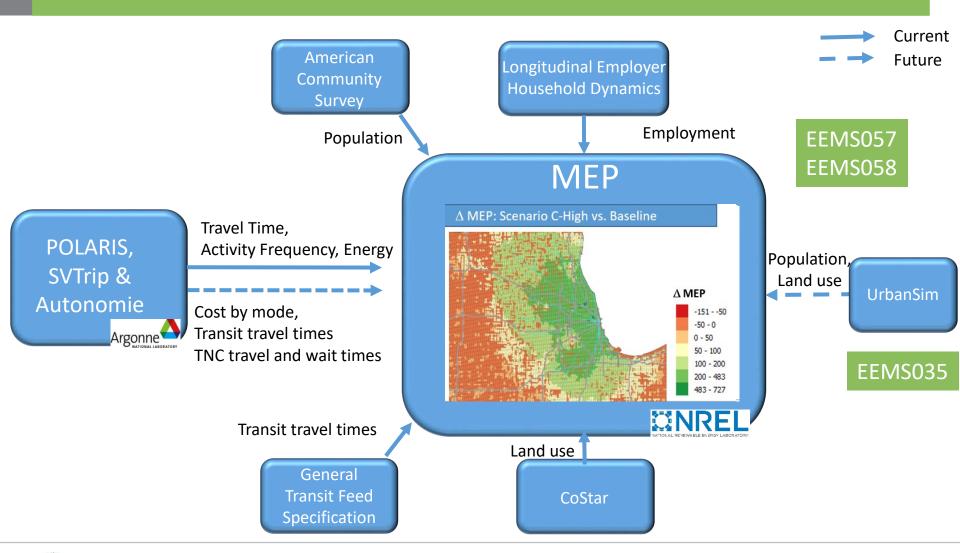








### AGGREGATING ALL RESULTS TO CALCULATE MEP





















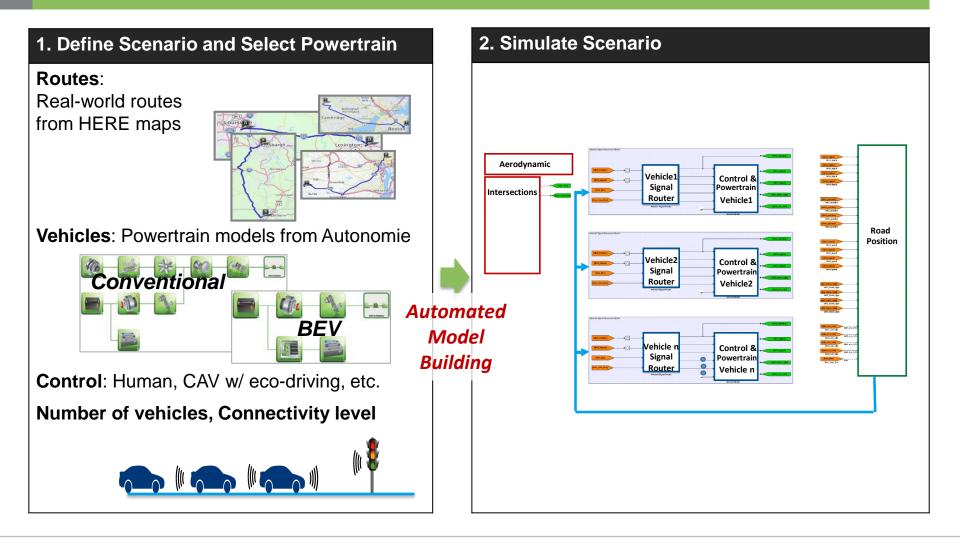








# AUTOMATED THE SIMULATION OF LARGE NUMBER OF SCENARIOS IN ROADRUNNER

















# HIGH PERFORMANCE COMPUTING (HPC) DEPLOYMENT LEVERAGES DOE R&D & ENABLES CALIBRATION

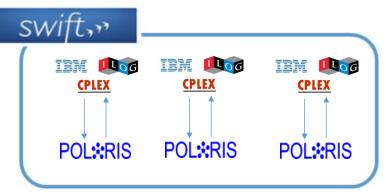
 HPC Computational/Optimization framework builds on Argonne's Swift/T and EMEWS<sup>(1)</sup> platforms to manage and run tens of thousands of simulations



Linux HPC
Distributed & Parallel HPC



Optimization (e.g. platooning, shared AVs...)





Windows HPC
Parallel computing

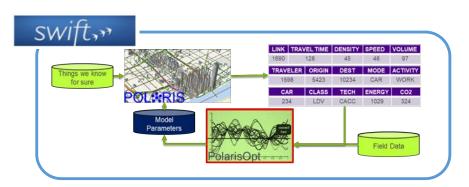


Windows HPC
Parallel computing





Calibration
(Critical for deployment & adoption)









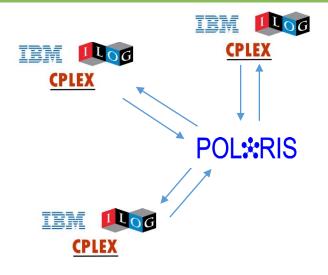




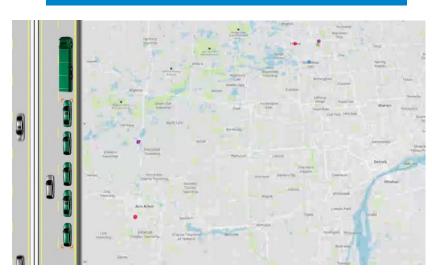


### **HPC ENABLES OPTIMIZATION & CONTROL**

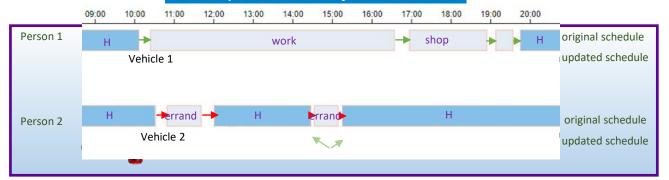
Implemented processes to efficiently link to external optimization tools



#### **Example: Platoon Formation Decision**



#### Example: Personally Owned AVs



Critical for "All About Me" scenario













# NEW WORKFLOWS DEVELOPED IN AMBER FOR DEPLOYMENT/ADOPTION

AMBER is a new workflow manager developed over the past 5 years

Accurate Vehicle Energy Consumption, Performance and Cost

Autonomie RoadRunner

AMBER

SVTrip POLARIS

Multi-Vehicle Simulation Environment for Control

Agent based transportation system simulation

Deployment critical
to support
Metropolitan
Planning
Organizations, Cities
and organizations
on Future Strategic
Plans

=> Identify solutions minimizing infrastructure investments

Stochastic Vehicle
Trip Profile
Prediction from
Geographic
Information
System (GIS)





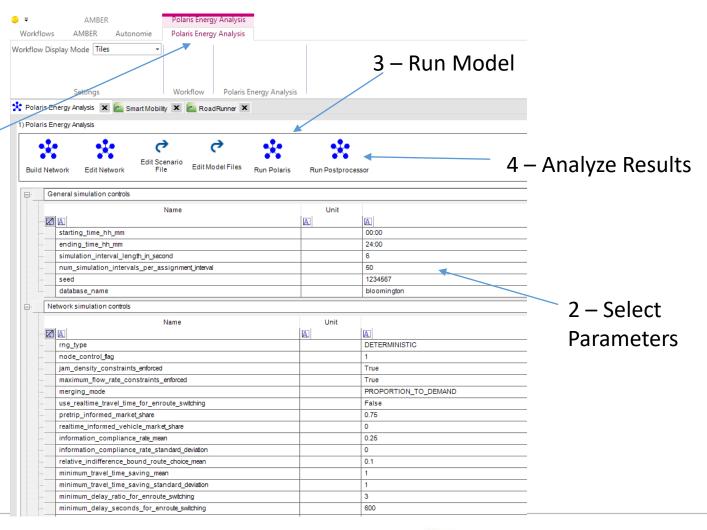






### **NEW AMBER WORKFLOW - POLARIS EXAMPLE**

1 – Load Existing POLARIS Model









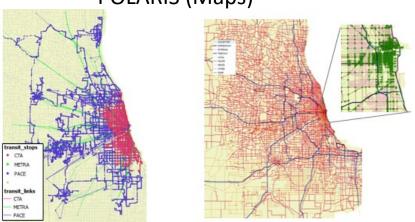




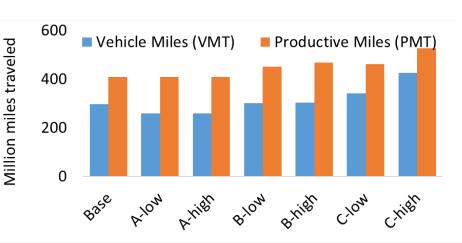


# **OUTPUT VISUALIZATION TOOLS STATIC**

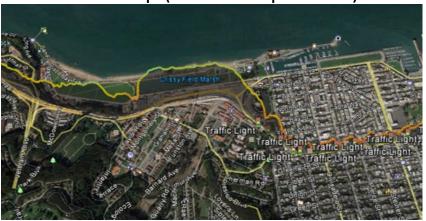
POLARIS (Maps)



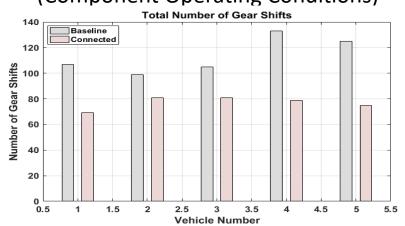
POLARIS (Results)



SVTrip (Vehicle Trip Profile)



RoadRunner (Component Operating Conditions)















# OUTPUT VISUALIZATION TOOLS DYNAMIC

### From entire metropolitan areas to individual vehicles

RoadRunner + CARLA

**POLARISGL** 































## **DETAILED CHICAGO MODEL INCLUDES TRANSPORTATION** INFRASTRUCTURE AND LAND USE

#### Transit network

- 35,077 nodes (CTA, PACE, METRA)
- 217,119 links (including auto network)
- 344 transit routes with 2,098 transit patterns
- 28,138 transit vehicle trips
- Intermodal and walking connections

## **Downtown Chicago** transit stops PACE transit\_links CTA METRA PACE

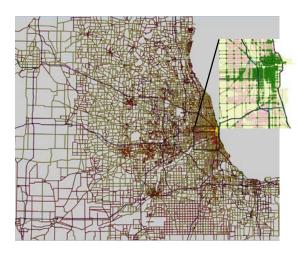
#### Street network

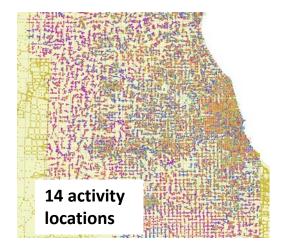
- 31,000 links with 18,900 nodes
- 7,900 traffic signals
- 12,500 stop signs
- 32.8 million trips (27 million by auto)

470,000 individual activity locations

Demand

- Associated with activity types
- Form start/end point for trips
- 270,000 parking locations with cost and capacity
- 10.2MM persons in 3.8MM HH











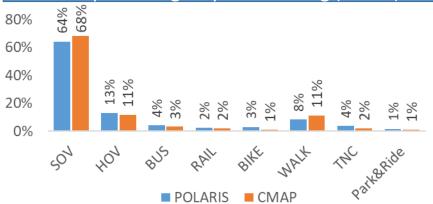




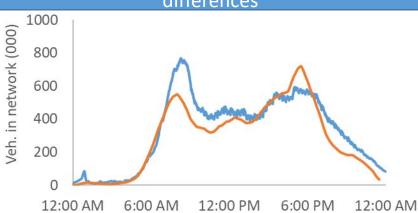


# CHICAGO BASELINE MODEL HAS BEEN CONTINUOUSLY CALIBRATED/VALIDATED SINCE 2012

# Mode shares closely matched to Chicago Metropolitan Agency for Planning (CMAP)



# In-network curves are very sensitive to model differences

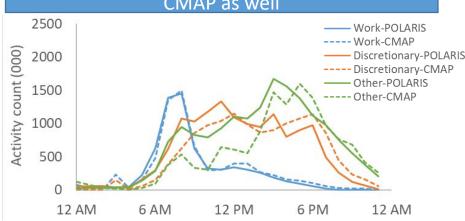


SOV - Single Occupancy Vehicle

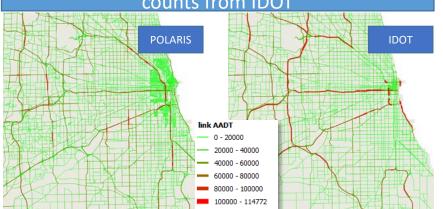
HOV - High Occupancy Vehicle

TNC – Transportation Network Company CMAP – Chicago Metropolitan Area Planning

# Activity counts & start times are similar to CMAP as well



# Simulated traffic counts compare closely to counts from IDOT







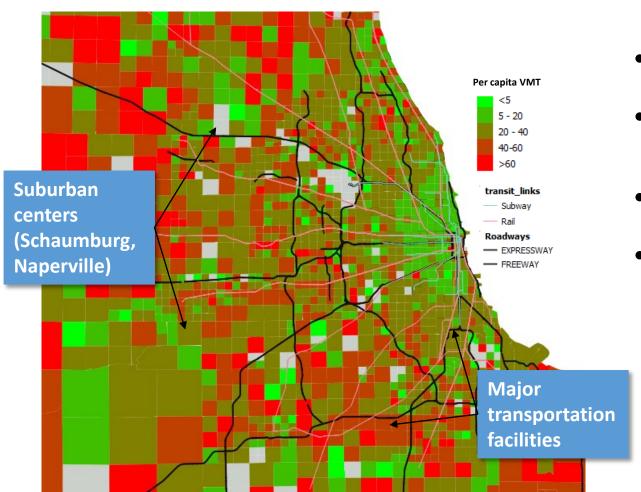








# VMT ON A PER CAPITA BASIS MUCH HIGHER IN SUBURBAN/RURAL AREAS AND AREAS WITH POOR ACCESSIBILITY



- VMT aggregated by home location of all travelers
- Much higher in areas away from major rail and interstate facilities
- Higher in areas further away from Chicago CBD
- There are multiple suburban pockets of lower average travel – polycentric Chicago Business Districts (CBDs)







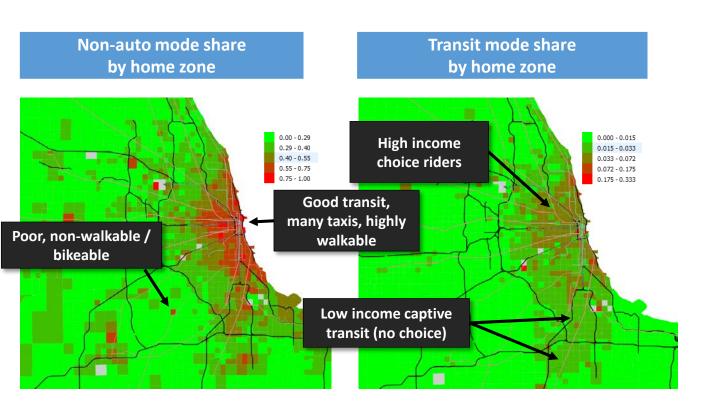


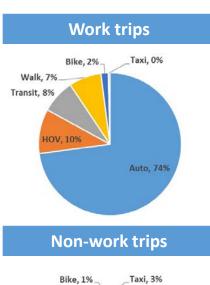


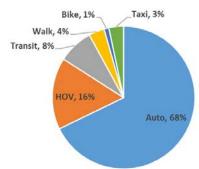


# MODE SHARES VARY SUBSTANTIALLY ACROSS THE REGION DEPENDING ON MANY FACTORS

Trip purpose, home accessibility, socio-demographics,...



















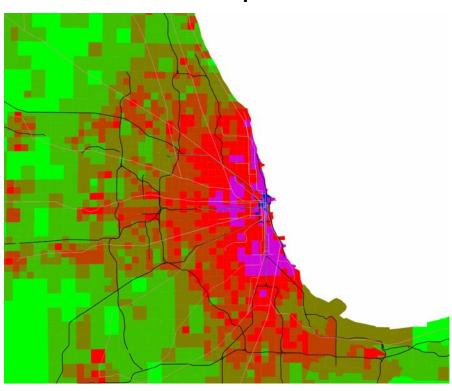
## TNC DISTRIBUTION

Pickup & Dropoffs concentrated downtown but still many occur in the suburbs

#### **Total TNC pickups**

# Pickup/dropoff per sq. km 0 - 20 20-50 50-100 100-200 200-600 600-2000 2000+

#### **Total TNC dropoffs**







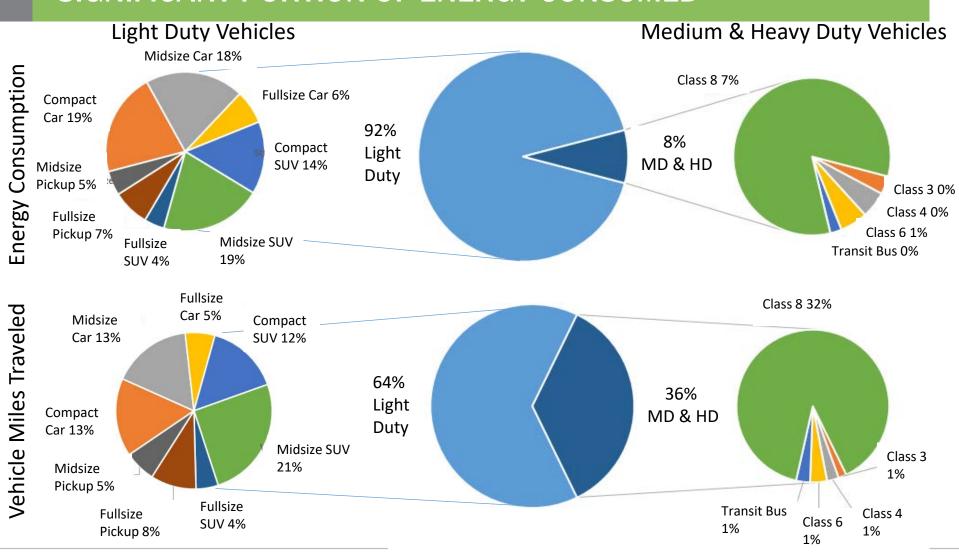








# MD/HD ACCOUNT FOR SMALL PORTION OF VMT BUT A SIGNIFICANT PORTION OF ENERGY CONSUMED





























### MULTIPLE SCENARIOS CONSIDERED (BASELINE + 3 FUTURES)

#### **Sharing is Caring (A)**



New technology (I.e., integrated Apps) enables people to significantly increase the use of transit, car sharing and multi-modal travel. Partial automation is being introduced mostly on the highway system.

#### **Technology Takes Over (B)**



Technology has taken over our lives, enabling a high usage of ride sharing and multi-modal trips as they are convenient affordable. As a result, private ownership has decreased, e-commerce is common as is telecommuting.

#### All About Me (C)





Fully automated vehicles within households are common with personal ownership resulting in low ride sharing market. The ability to own AVs leads to lower ecommerce and alternative work schedules, and feeds into urban sprawl.









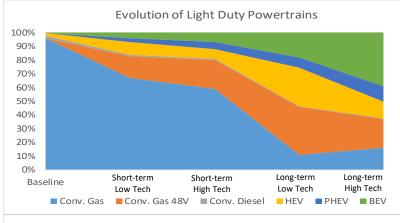


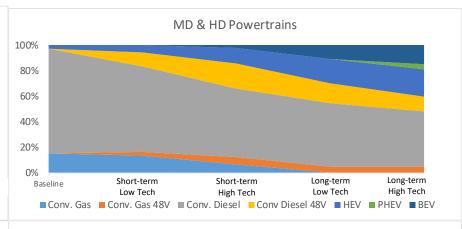


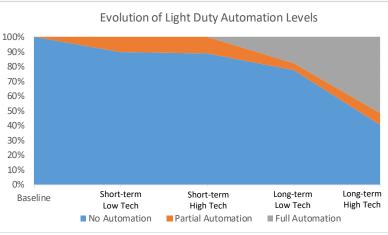
## **VEHICLE FLEET ASSUMPTIONS**

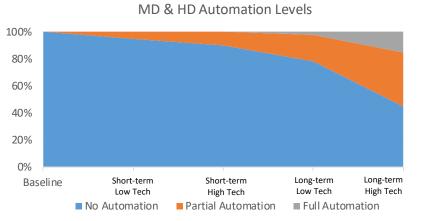
#### **Light Duty Vehicles**

#### Medium & Heavy Duty Vehicles













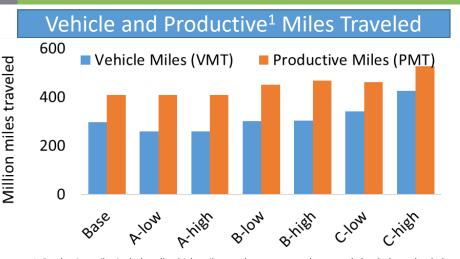


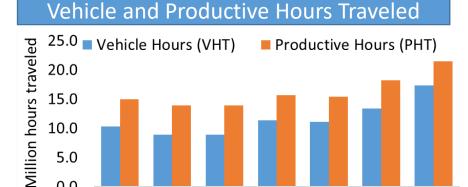




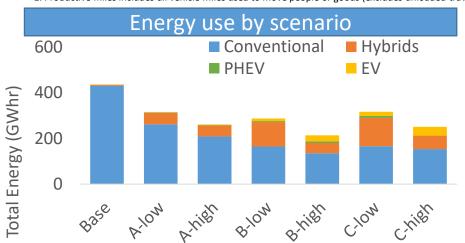


### POLARIS MODEL RESULTS: PRIVATE AV LESS EFFICIENT THAN SHARED FLEETS FOR REGIONAL ENERGY AND MOBILITY

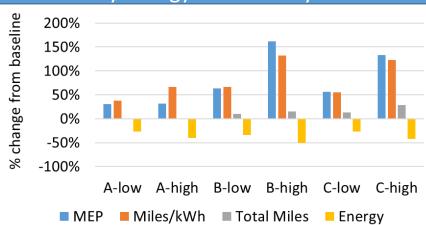




1. Productive miles includes all vehicle miles used to move people or goods (excludes unloaded travel miles)



#### **Mobility Energy Productivity metrics**





A - Sharing is caring

B - Technology takes over

C - All about me

Low - Vehicle business as usual **High – VTO Targets** Argonne



0.0



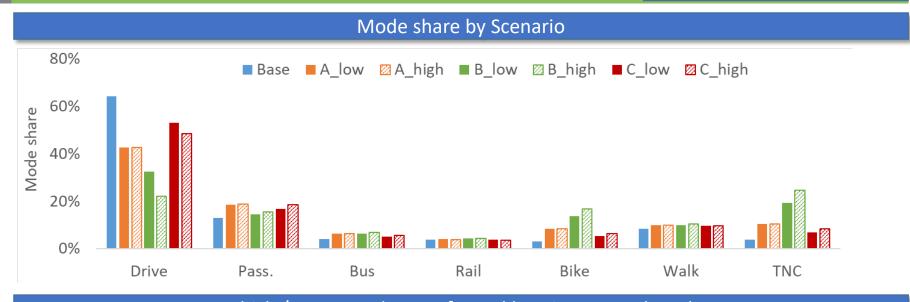




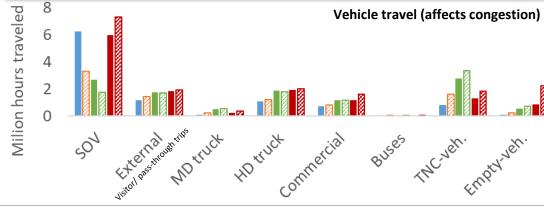


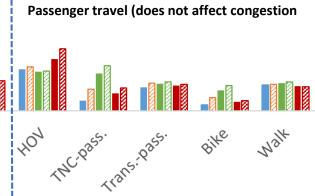
## CHANGES TO MOBILITY AND ENERGY ARE LARGELY DRIVE BY MODE SHIFTS AND SHIFT TO E-COMMERCE

**EEMS017, EEMS60, EEMS77** 



# Vehicle/Passenger hours of travel by trip type and mode





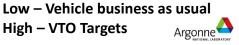


A - Sharing is caring

C - All about me

B - Technology takes over

**High – VTO Targets** 



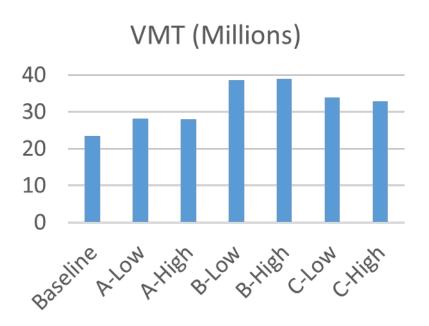




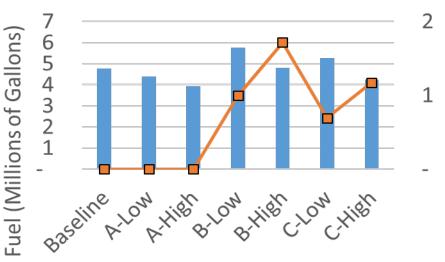




# E-COMMERCE, COMMODITY FLOWS DRIVE INCREASE IN TRUCK VMT IMPROVED TECHNOLOGY REQUIRED TO MITIGATE FUEL CONSUMPTION







FUEL -ELEC

# <u>Scenarios</u>—commodity flow growth:

A: Moderate

B & C: High

#### Scenarios—household delivery rate:

• Baseline: ~1 delivery per week

• A & B: 7 deliveries per week

• C: ~3-4 deliveries per week

EEMS060









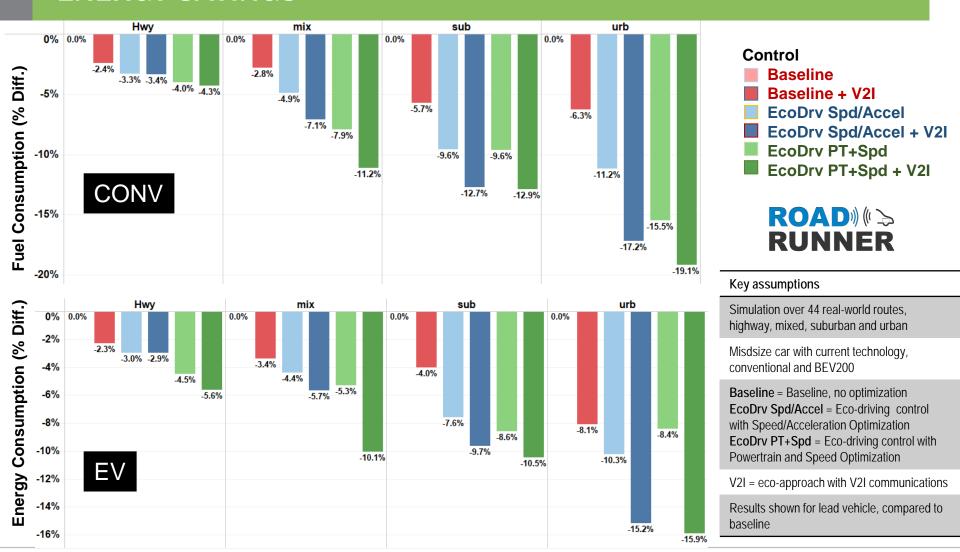






Electricity (Million kW-Hr)

# INDIVIDUAL CAV ECO-DRIVING VEHICLE CONTROL KEY TO ENERGY SAVINGS







EEMS016









### RESPONSE TO PREVIOUS YEAR REVIEWERS' COMMENTS

The project was not reviewed previously













### PARTNERSHIPS AND COLLABORATIONS



EEMS013, EEMS016, EEMS017, EEMS020, EEMS023, EEMS031, EEMS034, EEMS035, EEMS057, EEMS060, EEMS068, EEMS075, EEMS076, EEMS077, EEMS078, EEMS079



Improvement of CAV traffic flow model using CAV-specific fundamental diagrams



Shared Automated Vehicle (SAV) fleet modeling



Traveler behavior – Value of Travel Time



Activity scheduling and resource allocation



TNC modeling



Real-world vehicle energy use used to develop the Machine Learning Model













### REMAINING CHALLENGES AND BARRIERS

- Improve the implementation of each model interaction
  - -RoadRunner <-> Aimsun <-> POLARIS
  - -EVI-PRO <-> POLARIS
  - –UrbanSim <-> POLARIS
  - -Freight Demand <-> POLARIS
- When possible, further develop models to have similar level of fidelity, otherwise use translators
- Continue to validate the workflow with additional tools













### PROPOSED FUTURE RESEARCH

- Automate the process to support iterative simulations
- Implement and deploy processes with AMBER

Keep improving computational efficiency (HPC deployment,

convergence)

POL\*RIS



Linux OS

Distributed & Parallel HPC

Optimization / Control

No 3<sup>rd</sup> party license when possible



















## **SUMMARY - UNIQUE IMPLEMENTATION**

- Includes numerous partners (5 labs, 8 universities) each contributing unique expertise:
  - LBNL (micro-sim, on-road data, land use)
  - NREL (charging station location)
  - ORNL (parcel freight demand)
  - INL (EV charging)
  - LLNL (aerodynamics)
  - Univ. Calif Irvine (TNC repositioning)

George Mason Univ. (optimization/calibration)

- Texas A&M (CAV traffic flow model)
- Texas @ Austin (SAV fleet modeling)
- Rensselaer Polytechnic Inst. (freight)
- Univ. NewSouthWales (value of travel time)
- Washington State Univ. (TNC driver decision)

- Univ. Illinois@Chicago (activity scheduling and choice)

#### Smart Workflow

#### Integrated

- >10 partners
- > 12 tools
- VTO Benefit/Targets
- Includes economic impact
- Linkage with Life Cycle Analysis tools (GREET)

#### **High Fidelity**

- 100% agents simulated
- Representative vehicle models (VTO)
- Includes stop signs & traffic lights
- Enables vehicle speed dynamic
- Accurate energy consumption for each technology
- Component operating conditions

#### **Computationally Efficient**

- ~4h for 10M agents
- Entire process deployed with HPC



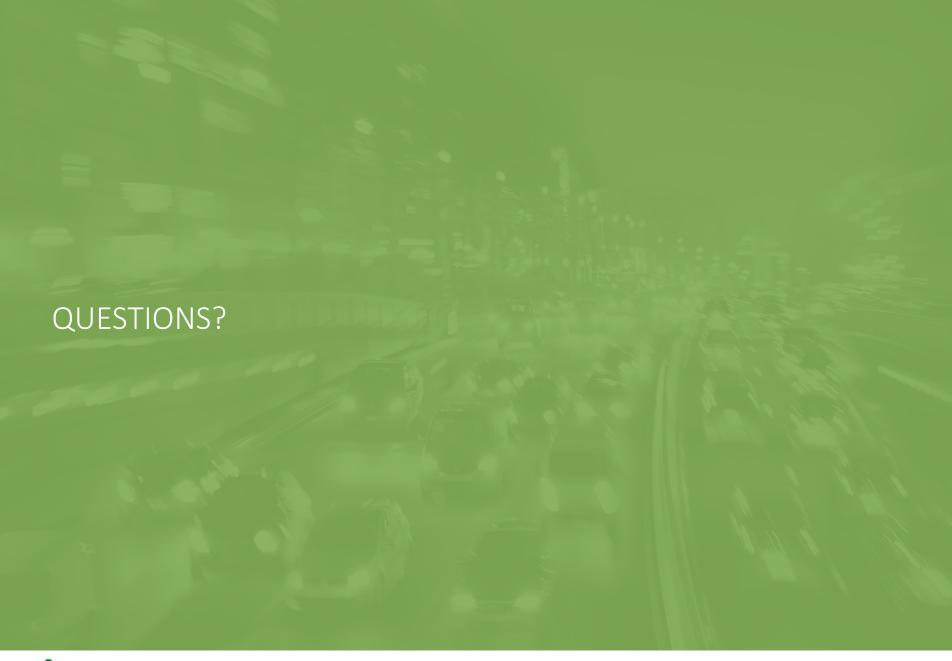






























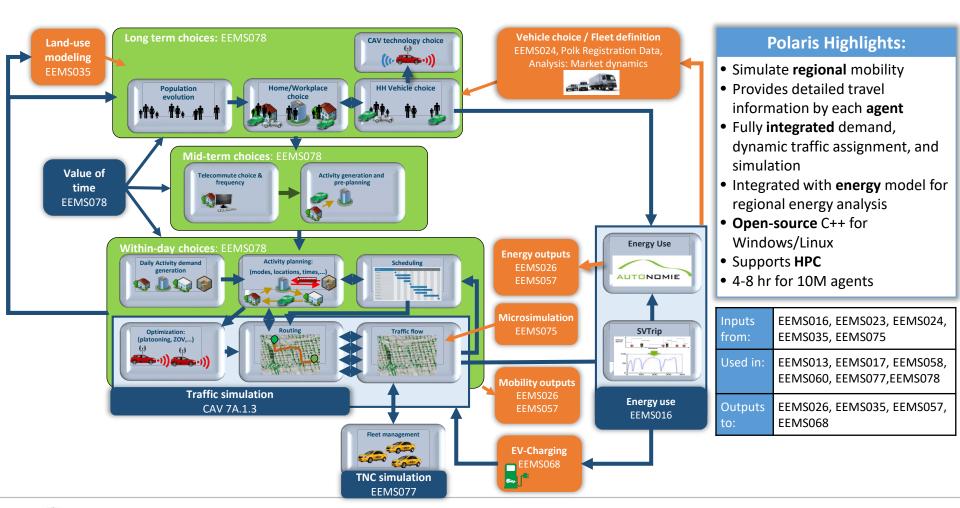








# POL\*RIS: AGENT-BASED ACTIVITY-TRAVEL SIMULATION MODEL SIMULATES REGIONAL MOBILITY











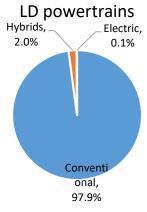




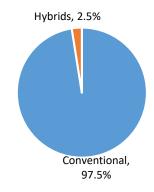
## **BASELINE FLEET COMPOSITION**

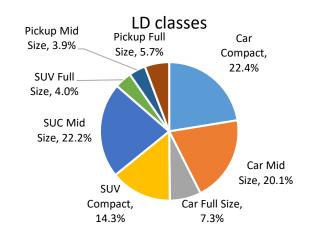
#### Consistent w/ POLK and IEA

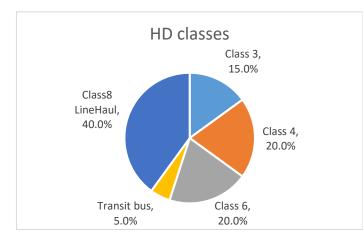
CAR_COMPACT-Gas-Conventional	21.2%
CAR_COMPACT-Gas-HEV	1.2%
CAR_MID_SIZE-Gas-Conventional	19.6%
CAR_MID_SIZE-Gas-HEV	0.5%
CAR_FULL_SIZE-Gas-Conventional	7.3%
SUV_COMPACT-Gas-Conventional	14.2%
SUV_COMPACT-Gas-HEV	0.1%
SUV_MID_SIZE-Gas-Conventional	22.0%
SUV_MID_SIZE-Gas-HEV	0.2%
SUV_FULL_SIZE-Gas-Conventional	4.0%
TRUCK_FULL_SIZE-Gas-Conventional	5.7%
TRUCK_MID_SIZE-Diesel-Conventiona	1.9%
TRUCK_MID_SIZE-Gas-Conventional	2.0%
Class3Box-Gas-Conventional	13.0%
Class3Shuttle-Gas-Conv	2.0%
Class4Delivery-Diesel-Conventional	20.0%
Class6P&D-Diesel-Conventional	20.0%
TransitBus-Diesel-Conventional	2.5%
TransitBus_Diesel-HEV	2.5%
Class8_LineHaul-Diesel-Conventional	40.0%

















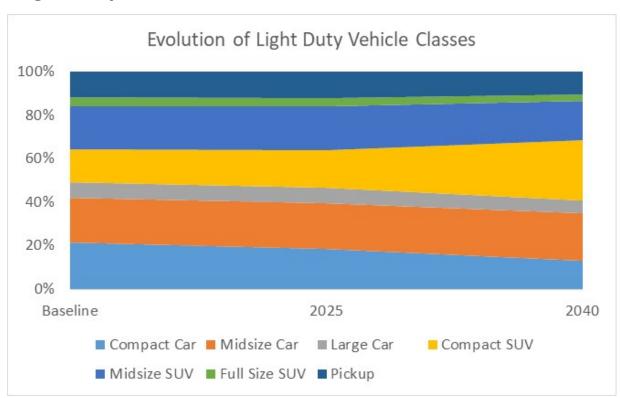






### **EVOLUTION OF VEHICLE CLASSES**

#### **Light Duty Vehicles**



Compact SUV market increase over time at the expense of passenger cars. Trend consistent with IEA

#### **Medium & Heavy Duty Vehicles**

Maintained current classes marked distribution constant







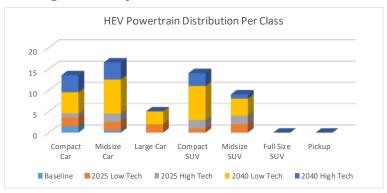






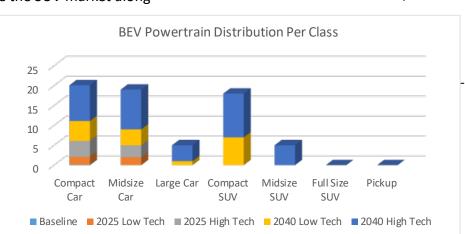
### **VEHICLE FLEET UPDATED**

#### **Light Duty Vehicles Electrification**



- HEVs expected to penetrate the SUV market along

with midsize car



PHEV Powertrain Distribution Per Class

14
12
10
8
6
4
2
0
Compact Midsize Large Car Compact SUV SUV SUV SUV

Baseline 2025 Low Tech 2025 High Tech 2040 Low Tech 2040 High Tech

- PHEVs predominant for midsize car

BEVs significant for compact car as well as SUVs













# PARAMETERS UPDATED/ADDED

#### Automation Distribution within Private and Fleet Light Duty Vehicles

Parameter not used for scenario	Variables	Baseline	(A) High sharing low automation	(B) High tech - mobility	(C)Low sharing high Automation
	Private Ownership	Current vehicle ownership based on POLK & current population by ZIP code	Low	Low	High
	Shared Use	1.3	1	1	1.3 (vehicle with driver) - 1.6 (vehicle without driver
	VOTT (Car mode only)	1	High (see table below)	Low (See table below)	Low (See table below)
	Propensity non-car modes	1	0.5	1	1
	Propensity telecommute	0.8 days per month	11.2 days per month	11.2 days per month	3.5 days per month
	E-Commerce	0.08 deliveries per person-day	0.5 deliveries per person-day	0.5 deliveries per person-day	0.2 deliveries per person-day
	Long Haul Commodity Flow	1% CAGR	1% CAGR	1.3% CAGR	1.3% CAGR
	Land use density	2017 Land Use	2017 Land Use	Long term planning (2050)	Urban sprawl
LDV privately owned	Non-Automated	98%	75% (low tech) / 74% (high tech)	41.5% (low tech) / 37.5% (high tech)	72.5% (low tech) / 35.5% (High tech)
	ownedL3/4	0%	5% (low tech) / 6% (high tech)	5% (Low Tech) / 8% (High tech)	5% (Low Tech) / 8% (High tech)
	L5	0%	0%	0%	12.5% (Low tech) / 41.5% (High tech)
LDV fleets (taxi, TNC)	Non-Automated	2%	15%	36% (low tech) / 3% (high tech)	5%
	L3/4	0%	5%	0%	0%
TIVE)	L5	0%	0%	17.5% (low tech) / 51.5% (high tech)	5% (low tech) / 10% (high tech)

New parameters added to improve consistency

Decided to use 2017 Land Use for all scenarios for AMR











